

THE BENEISH M-SCORE

Jeffrey Kercorian, Registered Investment Advisor • April 18, 2026

In 1998, a group of Cornell students ran Enron through an early version of the Beneish M-Score. The stock was trading at \$48. The model flagged it. Three years later, Enron collapsed in the largest accounting fraud in American history to that point. The people who ran the M-Score early got out. Everyone else lost everything.

That's what this model does. It reads financial statements the way a prosecutor reads a witness statement — not for what's there, but for what doesn't add up.

Who Is Messod Daniel Beneish

Beneish was born in Tunisia and came to the United States for graduate school, earning his MBA and PhD in Accounting from the University of Chicago in 1987. He started at Duke's Fuqua School of Business and has been at the Kelley School of Business at Indiana University Bloomington since 1993, where he holds the Alva L. Prickett Chair of Accounting.

He is not a celebrity academic. He is a working researcher who spent years studying earnings management, forensic accounting, and the specific patterns that appear in financial statements before fraud becomes public. His 1997 paper on GAAP violations showed that existing models missed manipulation in firms with extreme financial performance. He set out to build something better.

The result was his 1999 paper, "The Detection of Earnings Manipulation," published in the *Financial Analysts Journal*. He analyzed 74 firms cited by the SEC for manipulation against 2,332 non-manipulators from 1982 to 1992. He found patterns — declining gross margins, rising expenses, rapid sales growth accompanied by ballooning receivables — that appeared consistently before fraud became public. He turned those patterns into a weighted formula. He called the output the M-Score.

He has continued refining it. Studies in 2013 and 2020 expanded the model to predict economy-wide earnings management trends. His most recent work with Patrick Vorst applies aggregate M-Scores across thousands of firms to forecast recessions 5 to 8 quarters ahead. The model he built in 1999 to catch individual company fraud now functions as a macroeconomic early warning system.

How It Works

The M-Score is a probit regression model that takes eight financial ratios — each comparing the current year to the prior year — and combines them into a single number. Every ratio is looking for a specific kind of deterioration that tends to precede manipulated earnings.

The formula:

$$\begin{aligned}
M = & -4.84 + (0.92 \times \text{DSRI}) + (0.528 \times \text{GMI}) + (0.404 \times \text{AQI}) \\
& + (0.892 \times \text{SGI}) + (0.115 \times \text{DEPI}) - (0.172 \times \text{SGAI}) \\
& + (4.679 \times \text{TATA}) - (0.327 \times \text{LVGI})
\end{aligned}$$

A score above -1.78 suggests manipulation. Below -1.78, low risk. The model correctly identifies manipulators in 76% of cases.

Variable	What It Measures	Why It Matters
DSRI	Days Sales in Receivables Index	Customers suddenly taking longer to pay can mean fake sales booked to inflate revenue
GMI	Gross Margin Index	Shrinking margins with stable reported profits is a classic sign of hidden costs
AQI	Asset Quality Index	Rising intangible or deferred assets relative to total assets signals capitalized expenses
SGI	Sales Growth Index	Rapid revenue growth creates pressure and motive to manipulate
DEPI	Depreciation Index	Slowing depreciation artificially inflates earnings
SGAI	SG&A Expenses Index	Overhead growing faster than sales suggests operational deterioration being masked
TATA	Total Accruals to Total Assets	High non-cash earnings relative to assets is the single strongest fraud signal — weighted 4.679
LVGI	Leverage Index	Rising debt load increases pressure on management to hit earnings targets

TATA carries the heaviest weight in the formula by a wide margin — 4.679, more than five times the next highest. Non-cash earnings that don't translate into actual cash flow are the most reliable indicator that something is being manufactured. Cash is hard to fake. Accruals are not.

The model has limitations. It doesn't work well on banks and financial firms, where receivables and accruals behave differently by nature. It produces false positives in industries with structurally unusual financial profiles. It is a probabilistic tool, not a conviction. A high M-Score means look closer. It doesn't mean guilty.

The Cases

Enron

Cornell students ran an early version of the M-Score on Enron in 1998. The score flagged manipulation — inflated revenues, hidden debt structures, receivables that didn't match the sales story. The stock was at \$48. The model was right. Enron filed for bankruptcy in December 2001 in what was then the largest corporate collapse in American history. The people who paid attention to the early analysis avoided a catastrophic loss.

WorldCom and Tyco

Retrospective analyses of WorldCom and Tyco in the early 2000s showed elevated M-Scores in the years before their frauds became public. WorldCom had capitalized \$3.8 billion in operating expenses — exactly the kind of asset quality deterioration the AQI component detects. Tyco's manipulation was more diffuse but showed up in accruals and receivables patterns consistent with a high TATA.

Kangmei Pharmaceutical

Beneish himself consulted on the Kangmei Pharmaceutical case in China — one of the largest accounting frauds in Chinese capital market history. The company overstated cash and revenues by billions of renminbi over multiple years. The M-Score had flagged the pattern. The China Securities Regulatory Commission ultimately confirmed what the model had already suggested.

Short Selling

From 1993 to 2003, short-selling strategies based on the M-Score produced 14% annual hedged returns. The strategy is straightforward: short the companies with the highest M-Scores, go long the rest. The model's signal on who is likely manipulating earnings translates directly into a trading edge — not because the fraud is always confirmed, but because the risk-reward of betting against firms with deteriorating financial ratios is consistently favorable.

Bubbles and the Aggregate M-Score

Market bubbles and earnings fraud are not independent phenomena. They run together. When asset prices are rising on speculation rather than fundamentals, the pressure on management to match investor expectations intensifies. Weak oversight during booms makes manipulation easier to sustain. The South Sea Bubble of 1720, the 1920s stock boom, the dot-com era, the 2008 housing crisis — in every case, fraud rates rose with the bubble and the exposure of fraud accelerated the collapse.

Beneish's aggregate M-Score — applied across more than 2,000 firms simultaneously — picks up this pattern. Elevated aggregate scores appeared before the 2001 recession and again before 2008. In early 2023, aggregate M-Score levels matched those pre-recession peaks. The model built to catch individual company fraud now functions as a canary for the broader market.

The Magnificent Seven Problem

The S&P 500 is more concentrated than at any point in modern history. The Magnificent Seven — Apple, Microsoft, Nvidia, Amazon, Alphabet, Meta, Tesla — account for over 30% of the

index's total market capitalization as of mid-2024. The top 10 firms generate nearly 70% of economic profits in the index.

That concentration creates a specific systemic risk the M-Score is well positioned to monitor. If fraud or significant earnings manipulation emerges in one or two of those seven companies, the distortion doesn't stay contained — it moves the index. Running M-Scores on the Magnificent Seven is not paranoia. It is the logical application of a proven forensic tool to the stocks with the most leverage over the market as a whole.

Aggregate M-Scores rising across the broader market in a period of extreme concentration is the setup that preceded both 2001 and 2008. Whether 2023's elevated readings resolve the same way remains to be seen. The model doesn't predict timing. It identifies conditions.

Beneish built a tool in 1999 that caught Enron before Enron caught itself. He has spent 25 years refining it. The model correctly identifies manipulators 76% of the time, has generated 14% annual returns in short-selling applications, and now forecasts recessions. It works because it looks at the right things — cash versus accruals, receivables versus sales, margins versus reported profits — and it weights them based on what the data actually shows predicts fraud.

Every financial statement is a story someone is trying to tell. The M-Score asks whether the numbers behind the story add up.

Usually they do. When they don't, it's worth knowing early.

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Jeffrey Kercorian • Chief Investment Strategist • Wisdom Financial Advisors
jeff@wisdomfinancialadvisors.com • wisdomfinancialadvisors.com